

# 04\_modeling\_opec\_price\_forecasting

December 2, 2025

## 1 Modeling OPEC Sentiment vs PP Prices

Predict next-month PP\_EU using GPT comparison scores, hybrid index, FinBERT sentiment, and keyword densities.

```
[1]: from pathlib import Path
BASE_DIR = Path.cwd()
if BASE_DIR.name == 'notebooks':
    BASE_DIR = BASE_DIR.parent

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import mean_absolute_error
from sklearn.ensemble import RandomForestRegressor
from xgboost import XGBRegressor

sns.set_style('whitegrid')

data_path = BASE_DIR / 'data' / 'processed' / 'master_opec_price_model_dataset.
↪csv'
df = pd.read_csv(data_path)
df['date'] = pd.to_datetime(df['date'])
df = df.sort_values('date').reset_index(drop=True)
df.head()
```

```
[1]:
```

|   | date       | comparison_score | hybrid_index | finbert_sentiment | supply_up | \ |
|---|------------|------------------|--------------|-------------------|-----------|---|
| 0 | 2019-01-31 | 0.0              | 0.0          | 0.110262          | 0.004327  |   |
| 1 | 2019-02-28 | -0.7             | -0.7         | 0.046981          | 0.002033  |   |
| 2 | 2019-03-31 | -0.7             | -1.4         | 0.061729          | 0.008493  |   |
| 3 | 2019-04-30 | -0.6             | -2.0         | -0.239369         | 0.000517  |   |
| 4 | 2019-05-31 | -0.3             | -2.3         | -0.102873         | 0.003960  |   |

|   | supply_down | demand_up | demand_down | price_up | price_down | PP_EU  | Brent | \ |
|---|-------------|-----------|-------------|----------|------------|--------|-------|---|
| 0 | 0.004103    | 0.002233  | 0.004360    | 0.000000 | 0.001540   | 1385.0 | 59.77 |   |
| 1 | 0.004222    | 0.000220  | 0.004228    | 0.000097 | 0.001873   | 1385.0 | 63.63 |   |
| 2 | 0.003438    | 0.003655  | 0.004500    | 0.000073 | 0.001865   | 1410.0 | 66.66 |   |

|   |          |          |          |          |          |        |       |
|---|----------|----------|----------|----------|----------|--------|-------|
| 3 | 0.001615 | 0.000050 | 0.001818 | 0.000805 | 0.001073 | 1430.0 | 71.03 |
| 4 | 0.004860 | 0.002240 | 0.004933 | 0.000025 | 0.000148 | 1430.0 | 70.93 |

|   | WTI   | NatGas | PP_EU_next_month | Brent_next_month |
|---|-------|--------|------------------|------------------|
| 0 | 51.07 | 3.15   | 1385.0           | 63.63            |
| 1 | 54.53 | 2.69   | 1410.0           | 66.66            |
| 2 | 57.62 | 2.83   | 1430.0           | 71.03            |
| 3 | 63.22 | 2.62   | 1430.0           | 70.93            |
| 4 | 61.76 | 2.60   | 1430.0           | 63.35            |

## 1.1 Dataset overview

Columns include raw GPT comparison scores, cumulative hybrid index, FinBERT sentiment, keyword densities, and market prices aligned to month-end.

```
[2]: df.describe(include='all').T
```

```
[2]:
```

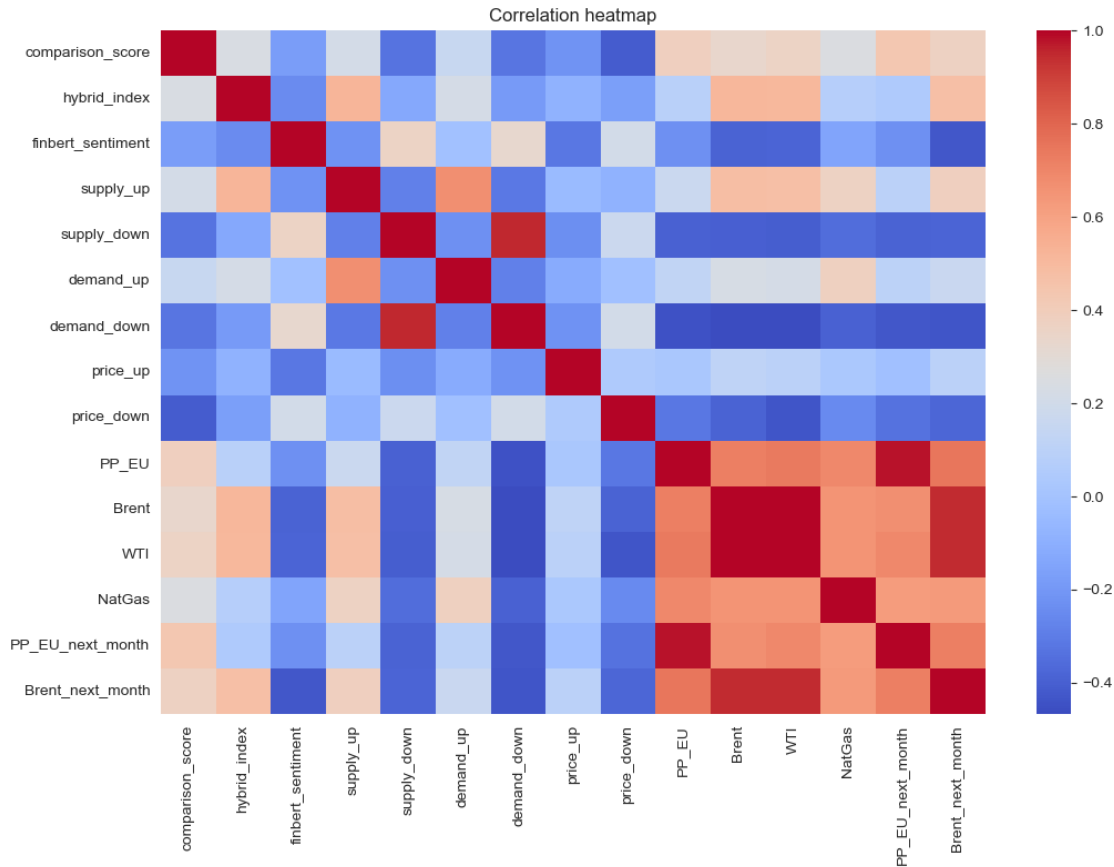
|                   | count | mean                | min \               |
|-------------------|-------|---------------------|---------------------|
| date              | 80    | 2022-05-26 19:12:00 | 2019-01-31 00:00:00 |
| comparison_score  | 80.0  | 0.11625             | -0.9                |
| hybrid_index      | 80.0  | -0.59625            | -10.8               |
| finbert_sentiment | 80.0  | -0.046661           | -0.245568           |
| supply_up         | 80.0  | 0.00598             | 0.000517            |
| supply_down       | 80.0  | 0.002217            | 0.0002              |
| demand_up         | 80.0  | 0.002424            | 0.00005             |
| demand_down       | 80.0  | 0.002306            | 0.000265            |
| price_up          | 80.0  | 0.000221            | 0.0                 |
| price_down        | 80.0  | 0.000401            | 0.0                 |
| PP_EU             | 80.0  | 1565.125            | 995.0               |
| Brent             | 80.0  | 72.775875           | 27.29               |
| WTI               | 80.0  | 68.38275            | 17.42               |
| NatGas            | 80.0  | 3.35625             | 1.71                |
| PP_EU_next_month  | 80.0  | 1565.875            | 995.0               |
| Brent_next_month  | 80.0  | 72.829875           | 27.29               |

|                   | 25%                 | 50% \               |
|-------------------|---------------------|---------------------|
| date              | 2020-09-22 12:00:00 | 2022-05-15 12:00:00 |
| comparison_score  | -0.45               | 0.35                |
| hybrid_index      | -5.525              | -0.75               |
| finbert_sentiment | -0.119719           | -0.057719           |
| supply_up         | 0.003641            | 0.006189            |
| supply_down       | 0.001404            | 0.001944            |
| demand_up         | 0.001324            | 0.002092            |
| demand_down       | 0.001469            | 0.001999            |
| price_up          | 0.000025            | 0.000046            |
| price_down        | 0.000048            | 0.000112            |
| PP_EU             | 1385.0              | 1451.25             |

|                  |         |         |
|------------------|---------|---------|
| Brent            | 63.855  | 73.555  |
| WTI              | 57.4775 | 70.375  |
| NatGas           | 2.3675  | 2.745   |
| PP_EU_next_month | 1385.0  | 1451.25 |
| Brent_next_month | 64.05   | 73.555  |

|                   | 75%                 | max                 | std        |
|-------------------|---------------------|---------------------|------------|
| date              | 2024-02-07 06:00:00 | 2025-09-30 00:00:00 | NaN        |
| comparison_score  | 0.6                 | 0.8                 | 0.563094   |
| hybrid_index      | 5.05                | 9.3                 | 5.840007   |
| finbert_sentiment | 0.027162            | 0.300448            | 0.114086   |
| supply_up         | 0.008089            | 0.012422            | 0.002935   |
| supply_down       | 0.002726            | 0.008483            | 0.001372   |
| demand_up         | 0.003445            | 0.00717             | 0.001556   |
| demand_down       | 0.002699            | 0.007567            | 0.00139    |
| price_up          | 0.000456            | 0.000958            | 0.000294   |
| price_down        | 0.00064             | 0.001873            | 0.000494   |
| PP_EU             | 1722.5              | 2475.0              | 376.825703 |
| Brent             | 83.27               | 118.14              | 18.021506  |
| WTI               | 78.865              | 115.19              | 18.177425  |
| NatGas            | 3.735               | 8.71                | 1.62051    |
| PP_EU_next_month  | 1722.5              | 2475.0              | 376.522248 |
| Brent_next_month  | 83.27               | 118.14              | 17.988483  |

```
[3]: plt.figure(figsize=(12, 8))
corr = df.corr(numeric_only=True)
sns.heatmap(corr, cmap='coolwarm', annot=False)
plt.title('Correlation heatmap')
plt.show()
```



## 1.2 Train/validation split

Use the earliest 80% of months for training and the latest 20% for validation to respect time order.

```
[4]: feature_cols = [
    'comparison_score', 'hybrid_index', 'finbert_sentiment',
    'supply_up', 'supply_down', 'demand_up', 'demand_down',
    'price_up', 'price_down', 'Brent', 'WTI', 'NatGas', 'PP_EU',
]

target = 'PP_EU_next_month'
split_idx = int(len(df) * 0.8)
X_train, X_val = df.loc[:split_idx - 1, feature_cols], df.loc[split_idx:,
    ↪feature_cols]
y_train, y_val = df.loc[:split_idx - 1, target], df.loc[split_idx:, target]
dates_val = df.loc[y_val.index, 'date']

naive_pred = df.loc[y_val.index, 'PP_EU']
naive_mae = mean_absolute_error(y_val, naive_pred)
print('Naive MAE (predict current PP as next month):', round(naive_mae, 2))
```

Naive MAE (predict current PP as next month): 11.25

```
[5]: rf = RandomForestRegressor(n_estimators=300, random_state=42,
    ↪min_samples_leaf=2)
rf.fit(X_train, y_train)
rf_pred = rf.predict(X_val)
rf_mae = mean_absolute_error(y_val, rf_pred)

xgb = XGBRegressor(
    random_state=42, n_estimators=400, learning_rate=0.05, max_depth=4,
    subsample=0.9, colsample_bytree=0.9
)
xgb.fit(X_train, y_train)
xgb_pred = xgb.predict(X_val)
xgb_mae = mean_absolute_error(y_val, xgb_pred)

results = pd.DataFrame(
    [
        ['Naive (current PP)', naive_mae],
        ['RandomForest', rf_mae],
        ['XGBoost', xgb_mae],
    ],
    columns=['model', 'mae']
)
results
```

```
[5]:
```

|   | model              | mae       |
|---|--------------------|-----------|
| 0 | Naive (current PP) | 11.250000 |
| 1 | RandomForest       | 23.632629 |
| 2 | XGBoost            | 37.491753 |

```
[6]: val_df = pd.DataFrame({
    'date': dates_val,
    'actual': y_val.values,
    'rf_pred': rf_pred,
    'xgb_pred': xgb_pred,
})

plt.figure(figsize=(12, 6))
plt.plot(val_df['date'], val_df['actual'], label='Actual', linewidth=2)
plt.plot(val_df['date'], val_df['rf_pred'], label='RF pred', alpha=0.8)
plt.plot(val_df['date'], val_df['xgb_pred'], label='XGB pred', alpha=0.8)
plt.legend()
plt.title('Validation: next-month PP_EU')
plt.xlabel('Date')
plt.ylabel('EUR/t')
plt.xticks(rotation=45)
```

```
plt.tight_layout()
plt.show()
```

